

# Accounting for Correlated Satellite Observation Error in NAVGEM

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*Presented at the AMS Annual Meeting*  
*4-8 January 2015*  
*Phoenix, AZ*



# Sources of Observation Error



- 1) Instrument error (usually, but not always, uncorrelated)
- 2) Mapping operator (H) error (interpolation, radiative transfer)
- 3) Pre-processing, quality control, and bias correction errors
- 4) Error of representation (sampling or scaling error), which can lead to correlated error:

True Temperature in Model Space  
Subgrid Scale Temperature

T=28°	T=38°	T=58°
T=30°	T=44°	T=61°
T=32°	T=53°	T=63°





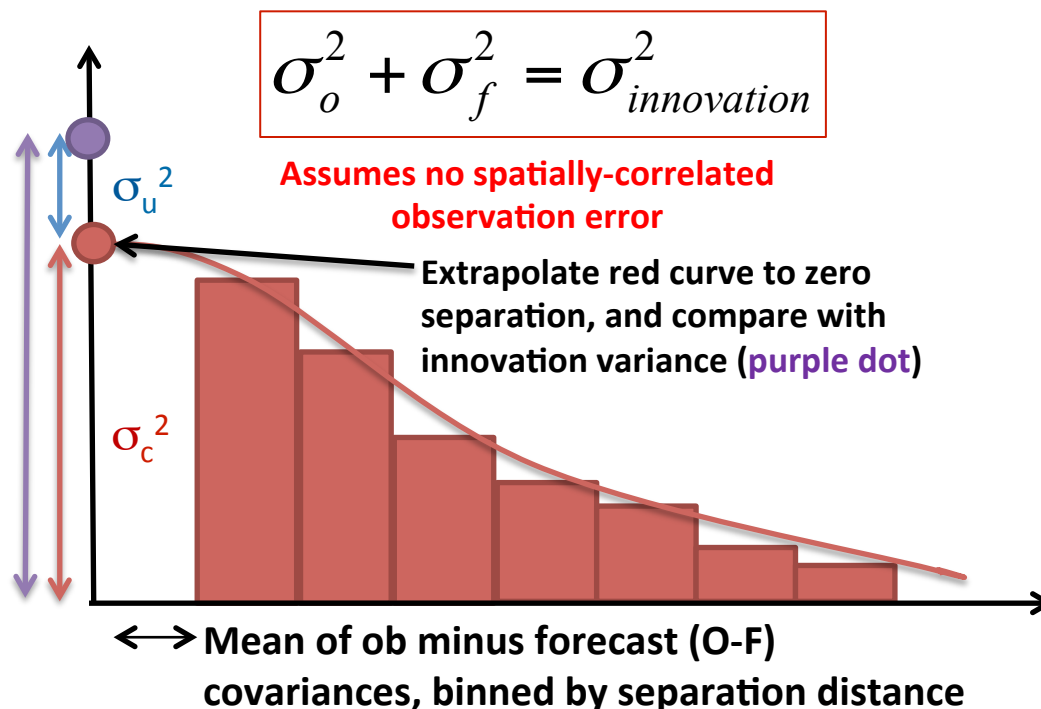
# Current State of the Art

- Most\* current DA methods generally assume no correlations between observations at different levels or locations (i.e., a diagonal **R**)
- To compensate for observation errors that *are* actually correlated, one or more of the following is typically done:
  - Discard (“**thin**”) observations until the remaining ones are uncorrelated (Bergman and Bonner (1976), Liu and Rabier (2003))
  - Local averaging (“**superobbing**”) (Berger and Forsythe (2004))
  - **Inflate** the observation error variances (Stewart et al. (2008, 2013))
- Recent theoretical studies (e.g. Stewart et al. (2008, 2013)) that include even *approximate* correlation structures use a diagonal **R** with variance inflation
- \*In January, 2013, the Met Office went operational with a vertical observation error covariance submatrix for the IASI instrument, which showed forecast benefit in seasonal testing in both hemispheres (Weston et al. (2014))



# Hollingsworth-Lönnberg Method

(Hollingsworth and Lönnberg, 1986)



- Use innovation statistics from a dense observing network
- Assume horizontally uncorrelated observation errors
- Calculate a histogram of background innovation covariances binned by horizontal separation
- Fit an isotropic correlation model, extrapolate to zero separation to estimate the correlated (forecast) and uncorrelated (observation) error partition



# Desroziers Method

(Desroziers *et al.* 2005)



- From O-F, O-A, and A-F statistics, the observation error covariance matrix **R**, the representer  $\mathbf{HBH}^T$ , and their sum can be diagnosed
- This method is sensitive to the **R** and  $\mathbf{HBH}^T$  that is prescribed in the DA system
- An iterative approach may be necessary

$$E \left[ \mathbf{d}_A^O \left( \mathbf{d}_F^O \right)^T \right] = \mathbf{R}$$

$$E \left[ \mathbf{d}_F^A \left( \mathbf{d}_F^O \right)^T \right] = \mathbf{HBH}^T$$

$$E \left[ \mathbf{d}_F^O \left( \mathbf{d}_F^O \right)^T \right] = \mathbf{R} + \mathbf{HBH}^T$$





# 4DVar Primal Formulation

$$\underline{w} \equiv \underline{x} - \underline{x}_f = BH^T \left( HBH^T + R \right)^{-1} \left( \underline{y} - H\underline{x}_f \right)$$

$$\left( B^{-1} + H^T R^{-1} H \right) \underline{w} = \left( B^{-1} + H^T R^{-1} H \right) BH^T \left( HBH^T + R \right)^{-1} \left( \underline{y} - H\underline{x}_f \right)$$

$$\left( B^{-1} + H^T R^{-1} H \right) \underline{w} = H^T R^{-1} \left( \underline{y} - H\underline{x}_f \right)$$

$$\underline{s} \equiv B^{1/2} \underline{w}$$

Scale by  $B^{-1/2}$

$$\underline{w} = B^{-1/2} \underline{s}$$

$$B^{-1/2} \left( B^{-1} + H^T R^{-1} H \right) \left( B^{-1/2} \underline{s} \right) = B^{-1/2} H^T R^{-1} \left( \underline{y} - H\underline{x}_f \right)$$

$$\left( I + B^{-1/2} H^T \underline{R}^{-1} H \right) B^{-1/2} \underline{s} = B^{-1/2} H^T \underline{R}^{-1} \left( \underline{y} - H\underline{x}_f \right)$$

4D-var iteration is on this problem -- We need to invert R!



# 4DVar Dual Formulation

- An advantage of the dual formulation is that correlated observation error can be implemented directly
- No matrix inverse is required, which lifts some restrictions on the feasible size of a non-diagonal  $R$
- In particular, implementing horizontally correlated observation error is significantly less challenging

$$\tilde{R}^{-1/2}(HBH^T + \tilde{R})\tilde{R}^{-1/2}(\tilde{R}^{1/2}\underline{z}) = \tilde{R}^{-1/2}(\underline{y} - H\underline{x}_b)$$

$$(\tilde{R}^{-1/2}HBH^T\tilde{R}^{-1/2} + I)\underline{w} = \tilde{R}^{-1/2}(\underline{y} - H\underline{x}_b)$$

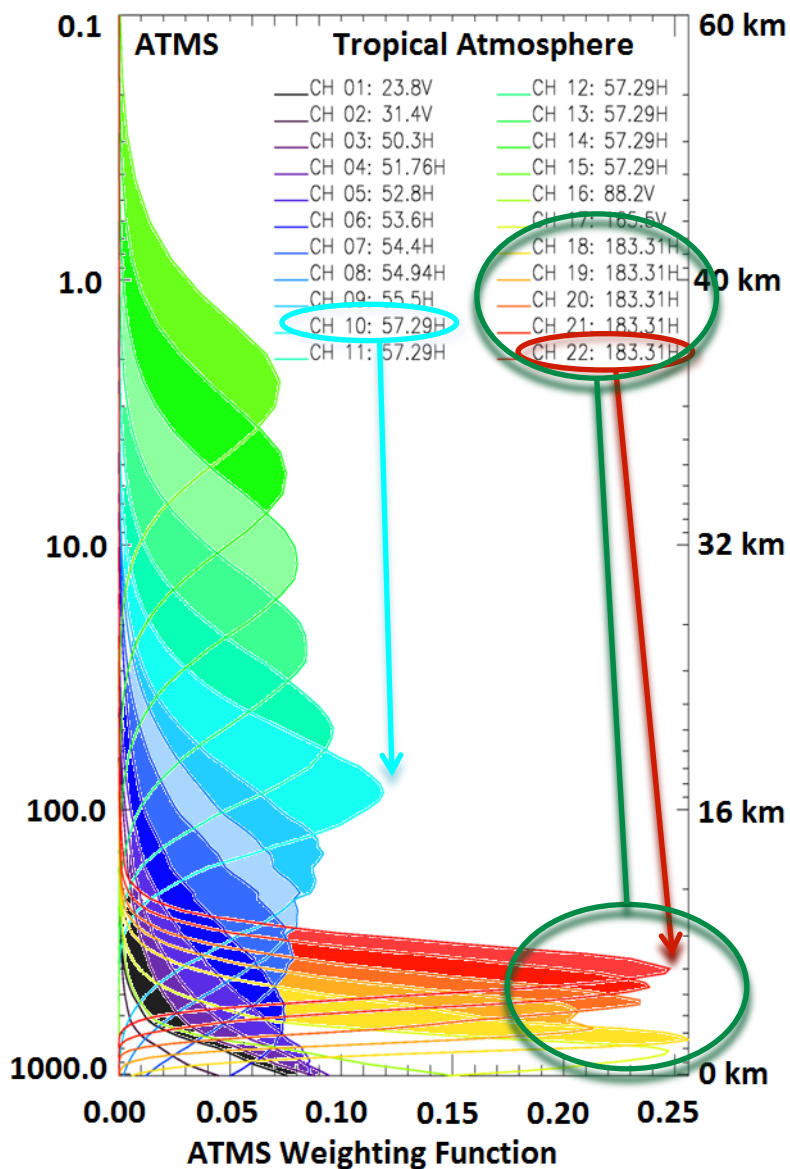
C

4D-Var iteration is on this problem – No need to invert

C!



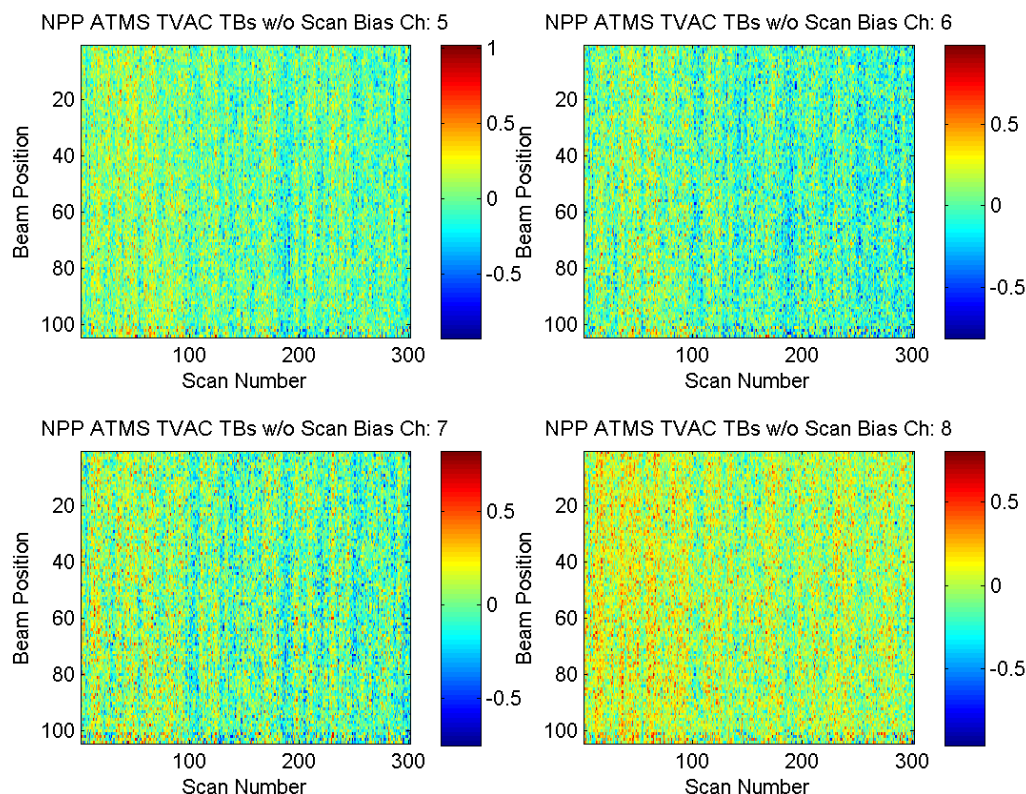
# Correlated Observation Error and the ATMS



## Advanced Technology Microwave Sounder (ATMS)

13 temperature channels

9 moisture channels



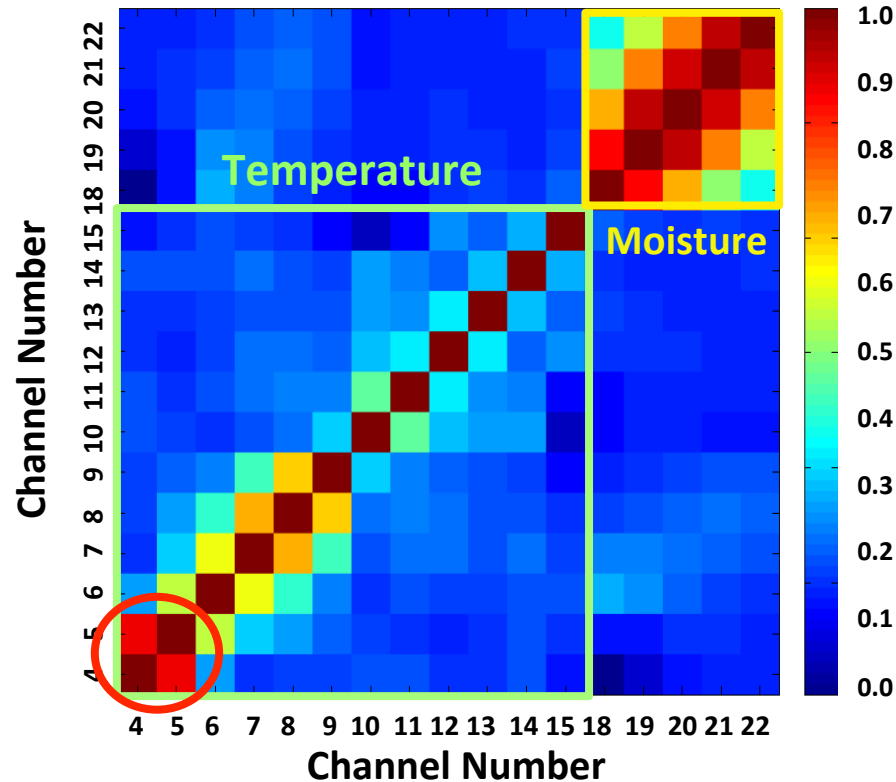




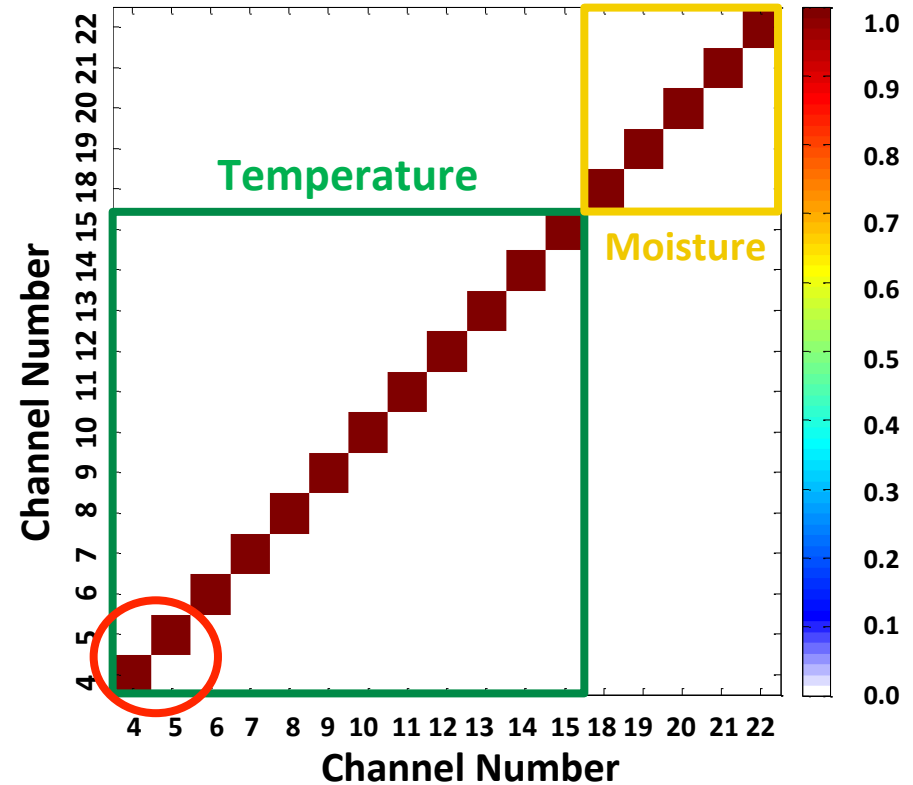
# Error Covariance Estimation for the ATMS



## Statistical Estimate



## Current Treatment



Estimating **R** is insufficient; we must be able to  
*use* it to assimilate data with correlated error



# Initial Experiments

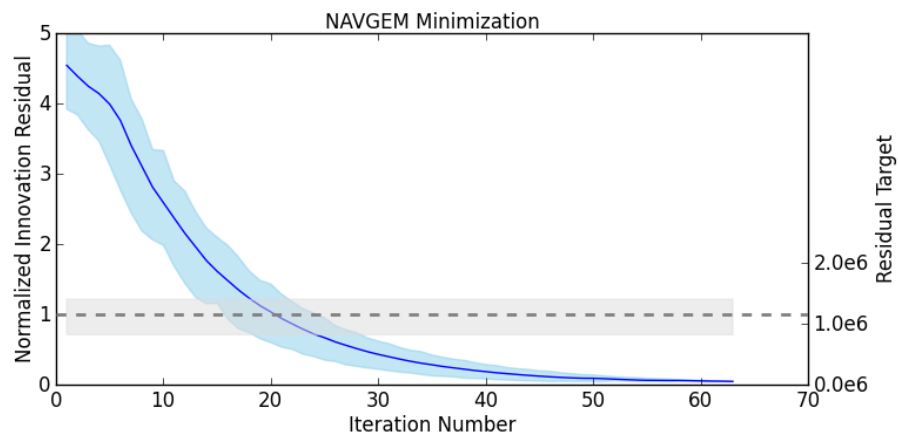
- We ran NAVGEM 1.3 at T425L60 resolution with the full suite of operational instruments for two months, from July 1, 2013 through Aug 31, 2013
- The control experiment used a diagonal **R** for the ATMS instrument
- The ATMS experiment used the **R** diagnosed from the Desroziers method applied to three months of innovation statistics
- **R** was symmetrized, but not otherwise altered



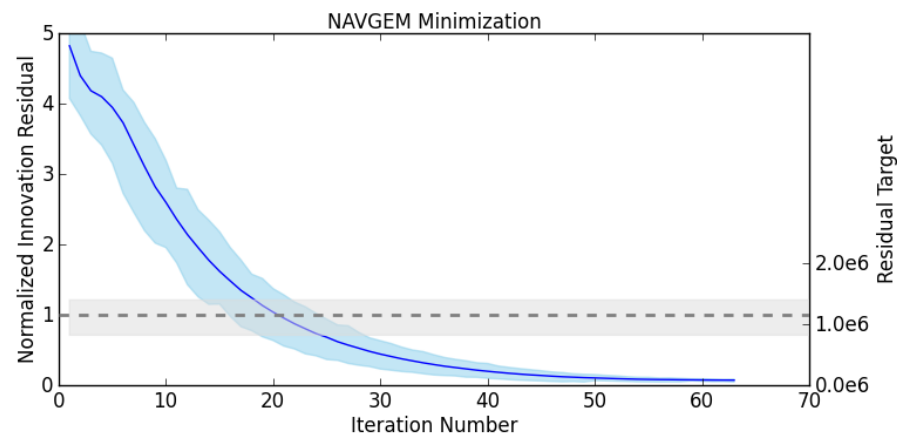
# Conjugate Gradient Convergence

2013080200

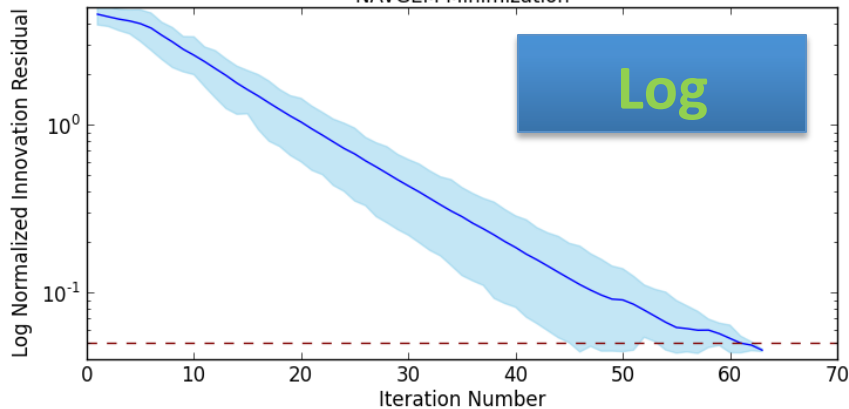
Control



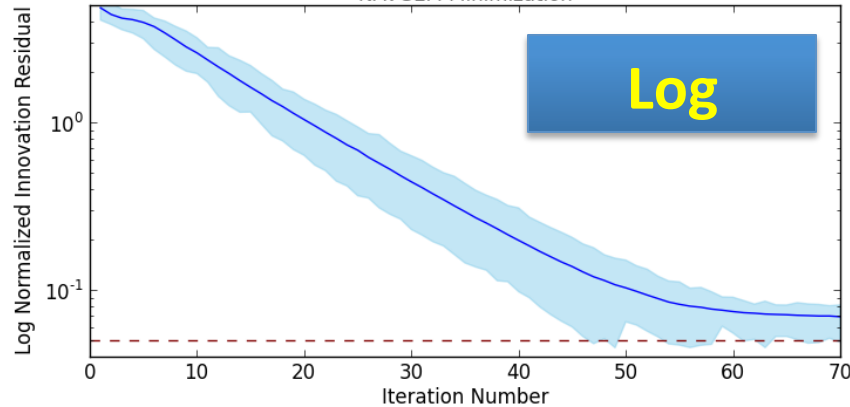
Experiment



NAVGEN Minimization



NAVGEN Minimization





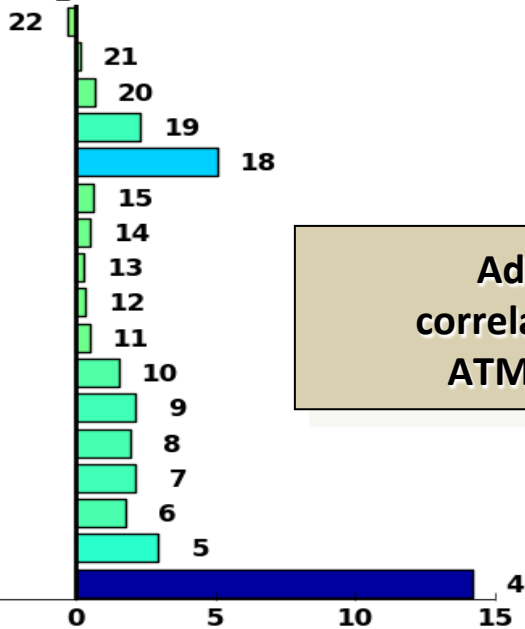
# Assess Forecast Impact with Observation Sensitivity Tools



Jul01-Aug02, 2013

NAVDAS-AR Observation Sensitivity

NPP\_ATMS

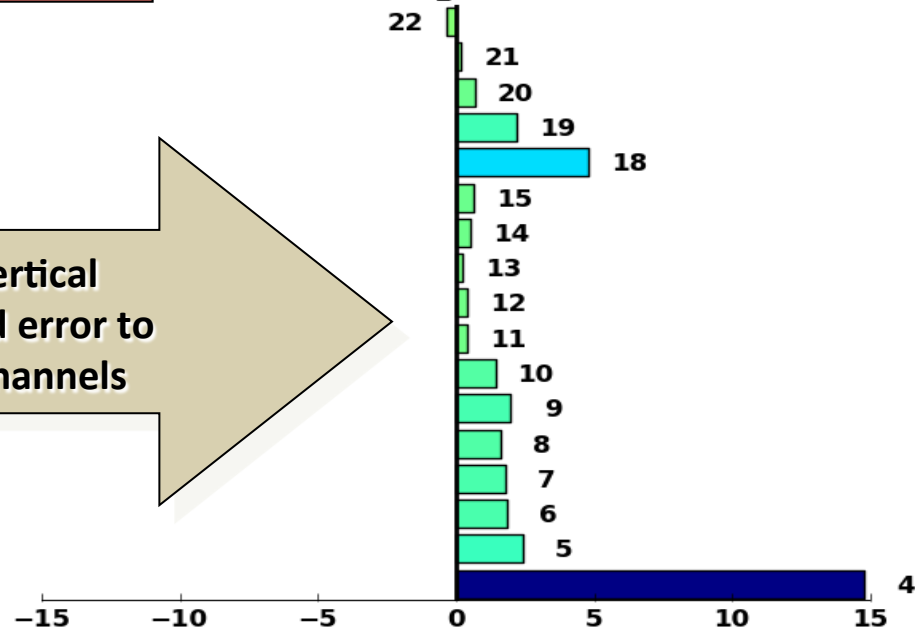


NAVDAS-AR  
Results

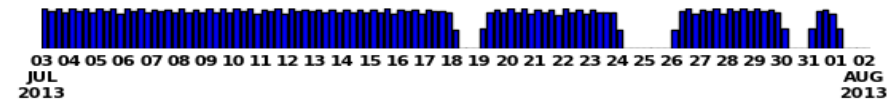
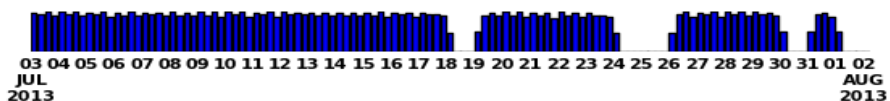
Jul01-Aug02, 2013

NAVDAS-AR Observation Sensitivity

NPP\_ATMS



Add vertical  
correlated error to  
ATMS channels



Adjoint-based system (Langland and Baker, 2004) enables **rapid assessment** of changes to the DA system





# Preliminary Results

## 2013070100-2013083012

Lead Time

0 24 48 72 96 120

atms by Lead Time 2013070100-2013083012

Show All entries

Reference	Level	Metric	Variable	Level type	Region	0	24	48	72	96	120	Mean
Fixed Buoy	None	Mean Error	Wind Speed	surface	Northern Hemisphere	😊	😊	😊	😊	😊	😊	+
Fixed Buoy	None	Mean Error	Wind Speed	surface	Southern Hemisphere	😊	😊	😊	😊	😊	😊	+
Fixed Buoy	None	Mean Error				😊	😊	😊	😊	😊	😊	+
Radiosondes	50.0	RMS Error				😊	😊	😊	😊	😊	😊	+
Radiosondes	250.0	RMS Error				😊	😊	😊	😊	😊	😊	+
Radiosondes	250.0	Vector RMS Error	Wind	pressure	Global	😊	😊	😊	😊	😊	😊	+
Radiosondes	500.0	RMS Error	Geopotential Height	pressure	Global	😊	😊	😊	😊	😊	😊	+
Radiosondes	850.0	RMS Error	Air Temperature	pressure	Global	😊	😊	😊	😊	😊	😊	+
Radiosondes	850.0	Vector RMS Error	Wind	pressure	Global	😊	😊	😊	😊	😊	😊	+
Self Analysis	100.0	RMS Error	Geopotential Height	pressure	Atlantic Region	😊	😊	😊	😊	😊	😊	+
Self Analysis	100.0	RMS Error	Geopotential Height	pressure	Eastern Pacific	😊	😊	😊	😊	😊	😊	+
Self Analysis	100.0	RMS Error				😊	😊	😊	😊	😊	😊	+
Self Analysis	100.0	RMS Error				😊	😊	😊	😊	😊	😊	+
Self Analysis	100.0	RMS Error	Geopotential Height	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	100.0	RMS Error	Geopotential Height	pressure	Western Pacific	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	RMS Error	Geopotential Height	pressure	Atlantic Region	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	RMS Error	Geopotential Height	pressure	Eastern Pacific	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	RMS Error	Geopotential Height	pressure	Northern Hemisphere	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	RMS Error	Geopotential Height	pressure	Southern Hemisphere	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	RMS Error	Air Temperature	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	RMS Error	Geopotential Height	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	RMS Error	Geopotential Height	pressure	Western Pacific	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	Vector RMS Error	Wind	pressure	Northern Hemisphere	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	Vector RMS Error	Wind	pressure	Southern Hemisphere	😊	😊	😊	😊	😊	😊	+
Self Analysis	200.0	Vector RMS Error	Wind	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Anomaly Correlation	Geopotential Height	pressure	Atlantic Region	😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Anomaly Correlation				😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Anomaly Correlation				😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Anomaly Correlation	Geopotential Height	pressure	Southern Hemisphere	😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Anomaly Correlation	Geopotential Height	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Anomaly Correlation	Geopotential Height	pressure	Western Pacific	😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	RMS Error	Air Temperature	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Vector RMS Error				😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Vector RMS Error				😊	😊	😊	😊	😊	😊	+
Self Analysis	500.0	Vector RMS Error				😊	😊	😊	😊	😊	😊	+
Self Analysis	850.0	RMS Error	Air Temperature	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	850.0	Vector RMS Error	Wind	pressure	Northern Hemisphere	😊	😊	😊	😊	😊	😊	+
Self Analysis	850.0	Vector RMS Error	Wind	pressure	Southern Hemisphere	😊	😊	😊	😊	😊	😊	+
Self Analysis	850.0	Vector RMS Error	Wind	pressure	Tropics	😊	😊	😊	😊	😊	😊	+
Self Analysis	1000.0	Anomaly Correlation				😊	😊	😊	😊	😊	😊	+
Self Analysis	1000.0	Anomaly Correlation				😊	😊	😊	😊	😊	😊	+
Self Analysis	1000.0	Anomaly Correlation				😊	😊	😊	😊	😊	😊	+
Self Analysis	1000.0	Anomaly Correlation				😊	😊	😊	😊	😊	😊	+
Self Analysis	1000.0	Anomaly Correlation				😊	😊	😊	😊	😊	😊	+

Global RMS Temperature Error at 250 hPa vs. Radiosondes

Global RMS Temperature Error at 850 hPa vs. Radiosondes

Tropical RMS Temperature Error at 500 hPa vs. Self-analysis

(NH/SH) RMS Vector Wind Error at 850 hPa vs. Self-analysis

Tropical RMS Vector Wind Error at 850 hPa vs. Self-analysis

NH Geopotential Anomaly Correlation at 1000 hPa vs. Self-analysis



# Conclusions

- The Hollingsworth-Lonnberg and Desroziers error covariance estimation methods can quantify correlated observation error
- The NAVGEM system allows for direct use of a non-diagonal **R**; implementing vertically correlated error is straightforward.
- Correctly accounting for correlated observation error in data assimilation may yield superior forecast results without a large computational cost



# Discussion

- How can we best estimate errors in Desroziers/Hollingsworth- Lönnberg diagnostics?
  - Should we expect agreement between different methods?
  - Will the Desroziers diagnostic converge if both R and B are incorrectly specified?
  - Amount of data required to estimate covariances? Seasonal dependence?
  - Best methods to symmetrize the Desroziers matrix?
- How to gauge improvement?
  - Do we also need to adjust to see overall improvement to the system?
  - How do we maintain the correct ratio for DA?
- What about convergence?
  - Should we do an eigenvalue scaling to improve the condition number?